

# Brute force KNN model

In [0]:

In [1]:

```
# Load the Drive helper and mount
from google.colab import drive

# This will prompt for authorization.
drive.mount('/content/drive')
```

Go to this URL in a browser: [https://accounts.google.com/o/oauth2/auth?client\\_id=947318989803-6bn6qk8qdgf4n4g3pfee6491hc0brc4i.apps.googleusercontent.com&redirect\\_uri=urn%3Aietf%3Awg%3Aoauth%3A2.0%b&scope=email%20https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdocs.test%20https%3A%2F%2Fwww.googleapis.com%2Fdrive.photos.readonly%20https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fpeopleapi.readonly&response\\_type=code](https://accounts.google.com/o/oauth2/auth?client_id=947318989803-6bn6qk8qdgf4n4g3pfee6491hc0brc4i.apps.googleusercontent.com&redirect_uri=urn%3Aietf%3Awg%3Aoauth%3A2.0%b&scope=email%20https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdocs.test%20https%3A%2F%2Fwww.googleapis.com%2Fdrive.photos.readonly%20https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fpeopleapi.readonly&response_type=code)

Enter your authorization code:  
.....

Mounted at /content/drive

In [0]:

```
import warnings
warnings.filterwarnings("ignore")

import sqlite3
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.feature_extraction.text import TfidfVectorizer

from sklearn.feature_extraction.text import CountVectorizer
from sklearn.metrics import confusion_matrix
from sklearn import metrics
from sklearn.metrics import roc_curve, auc
from nltk.stem.porter import PorterStemmer

import re
# Tutorial about Python regular expressions: https://pymotw.com/2/re/
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer

from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle

# using the SQLite Table to read data.
con = sqlite3.connect('database.sqlite')

#filtering only positive and negative reviews i.e.
# not taking into consideration those reviews with Score=3
filtered_data = pd.read_sql_query(""" SELECT * FROM Reviews WHERE Score != 3 """, con)

# Give reviews with Score>3 a positive rating, and reviews with a score<3 a negative rating.
def partition(x):
    if x < 3:
        return 0
    return 1

#changing reviews with score less than 3 to be positive and vice-versa
```

```

#changing reviews with score less than 5 to be positive and vice versa
actualScore = filtered_data['Score']
positiveNegative = actualScore.map(partition)
filtered_data['Score'] = positiveNegative
print(filtered_data.shape)

```

In [0]:

```

sorted_data=filtered_data.sort_values('ProductId', axis=0, ascending=True, inplace=False, kind='quicksort', na_position='last')
final=sorted_data.drop_duplicates(subset={"UserId","ProfileName","Time","Text"}, keep='first', inplace=False)
final.shape

```

In [0]:

```

final.head(2)

```

In [0]:

```

final=final[final.HelpfulnessNumerator<=final.HelpfulnessDenominator]
#Before starting the next phase of preprocessing lets see the number of entries left
print(final.shape)

#How many positive and negative reviews are present in our dataset?
final['Score'].value_counts()

```

In [0]:

```

final.sort_values('Time',axis=0,ascending=True,inplace=True,kind='quicksort')
final.head(2)

```

In [0]:

```

stop = set(stopwords.words('english')) #set of stopwords
sno = nltk.stem.SnowballStemmer('english') #initialising the snowball stemmer

def cleanhtml(sentence): #function to clean the word of any html-tags
    cleanr = re.compile('<.*?>')
    cleantext = re.sub(cleanr, ' ', sentence)
    return cleantext

def cleanpunc(sentence): #function to clean the word of any punctuation or special characters
    cleaned = re.sub(r'[?|!|\'|\"|#]',r'',sentence)
    cleaned = re.sub(r'[.,|)|(|\\|/]',r'',cleaned)
    return cleaned

```

In [0]:

```

#Code for implementing step-by-step the checks mentioned in the pre-processing phase
# this code takes a while to run as it needs to run on 500k sentences.
i=0
str1=' '
final_string=[]
all_positive_words=[] # store words from +ve reviews here
all_negative_words=[] # store words from -ve reviews here.
s=''
for sent in final['Text'].values:
    filtered_sentence=[]
    #print(sent);
    sent=cleanhtml(sent) # remove HTML tags
    for w in sent.split():
        for cleaned_words in cleanpunc(w).split():
            if((cleaned_words.isalpha()) & (len(cleaned_words)>2)):
                if(cleaned_words.lower() not in stop):
                    s=(sno.stem(cleaned_words.lower())).encode('utf8')
                    filtered_sentence.append(s)
                    if (final['Score'].values)[i] == 1:
                        all_positive_words.append(s) #list of all words used to describe positive reviews
                    if(final['Score'].values)[i] == 0:
                        all_negative_words.append(s) #list of all words used to describe negative reviews

```

```
VIEWED REVIEWS
        else:
            continue
    else:
        continue
    #print(filtered_sentence)
    str1 = b" ".join(filtered_sentence) #final string of cleaned words
    #print("*****")

    final_string.append(str1)
    i+=1
```

In [0]:

```
final['CleanedText']=final_string #adding a column of CleanedText which displays the data after pr
e-processing of the review
final['CleanedText']=final['CleanedText'].str.decode("utf-8")
final.head(3)
```

In [0]:

```
import pickle
pickle.dump(final, open('drive/My Drive/Colab Notebooks/knn/final.p', 'wb'))
#final_sent = pickle.load(open('data.p', 'rb'))
final.shape
```

In [1]:

```
import pickle
final = pickle.load(open('drive/My Drive/Colab Notebooks/gbdt/final.p', 'rb'))

from sklearn.model_selection import train_test_split
##Sorting data according to Time in ascending order for Time Based Splitting
time_sorted_data = final.sort_values('Time', axis=0, ascending=True, inplace=False, kind='quicksort',
na_position='last')
final.head(2)
```

Out[1]:

	Id	ProductId	UserId	ProfileName	HelpfulnessNumerator	HelpfulnessDenominator	Score	Time	Summ
138706	150524	0006641040	ACITT7DI6IDDL	shari zychinski	0	0	1	939340800	EV bo educati

138683	150501	0006641040	AJ46FKXOVC7NR	Nicholas A Mesiano	2	2	1	940809600	This w seri great to sp time
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In [2]:

```
# sampling the data for ease
y = final['Score'].iloc[0:40000]
x = final['CleanedText'].iloc[0:40000]
x.shape,y.shape
```

Out[2]:

((40000,), (40000,))

## BOW

In [3]:

```
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
```

```
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer
```

```
X_tra, X_tes, y_train, y_test = train_test_split(x,y,test_size=0.3,shuffle=False,random_state=0)
```

```
#Implementing BAG of words
```

```
bow = CountVectorizer()
```

```
X_tf =bow.fit_transform(X_tra)
```

```
# Standerdising the data
```

```
norm = StandardScaler(with_mean = False)
```

```
X_train = norm.fit_transform(X_tf)
```

```
# tfidf test
```

```
X_tfte = bow.transform(X_tes)
```

```
# Standerdising the data
```

```
X_test = norm.transform(X_tfte)
```

```
/usr/local/lib/python3.6/dist-packages/sklearn/utils/validation.py:475: DataConversionWarning: Data with input dtype int64 was converted to float64 by StandardScaler.
warnings.warn(msg, DataConversionWarning)
```

In [7]:

```
from sklearn.neighbors import KNeighborsClassifier
```

```
from sklearn.cross_validation import cross_val_score
```

```
from sklearn import linear_model
```

```
from sklearn.metrics import make_scorer
```

```
from sklearn.metrics import f1_score
```

```
# 10 fold Cross Validation
```

```
myList = list(range(0,15))
```

```
neighbors = list(filter(lambda x: x % 2 != 0, myList))# Finding prime numbers
```

```
cv_scores = []
```

```
for k in neighbors:
```

```
    knn = KNeighborsClassifier(n_neighbors=k,algorithm='brute')
```

```
    scores = cross_val_score(knn, X_train,y_train, cv=10, scoring='f1_weighted')
```

```
    print('\nThe cross validation score for K = {} is {}'.format(k,scores.mean()))
```

```
    cv_scores.append(scores.mean())
```

```
MSE = [1 - x for x in cv_scores]
```

```
optimal_k = neighbors[MSE.index(min(MSE))]
```

```
/usr/local/lib/python3.6/dist-packages/sklearn/cross_validation.py:41: DeprecationWarning: This module was deprecated in version 0.18 in favor of the model_selection module into which all the refactored classes and functions are moved. Also note that the interface of the new CV iterators are different from that of this module. This module will be removed in 0.20.
```

```
"This module will be removed in 0.20.", DeprecationWarning)
```

```
The cross validation score for K = 1 is 0.8409052274519677.
```

```
The cross validation score for K = 3 is 0.854205242845245.
```

```
The cross validation score for K = 5 is 0.8489215818144903.
```

```
The cross validation score for K = 7 is 0.8447328697023988.
```

```
The cross validation score for K = 9 is 0.8422545549630334.
```

```
The cross validation score for K = 11 is 0.8410442472820181.
```

```
The cross validation score for K = 13 is 0.8402215806876621.
```

In [8]:

```
print('\nThe optimal number of neighbors is %d.' % optimal_k)
```

```
The optimal number of neighbors is 3.
```

In [0]:

```
from sklearn.neighbors import KNeighborsClassifier
knn_optimal = KNeighborsClassifier(n_neighbors=3,algorithm='brute')

knn_optimal.fit(X_train,y_train)

pred = knn_optimal.predict(X_test)
```

In [11]:

```
from sklearn.metrics import f1_score,precision_score,recall_score
f1 = f1_score(y_test, pred, average='micro')
f1_per= f1*100

print('\nThe f1 score of the knn classifier of BOW for k = %d is %f%%' % (3, f1_per))
```

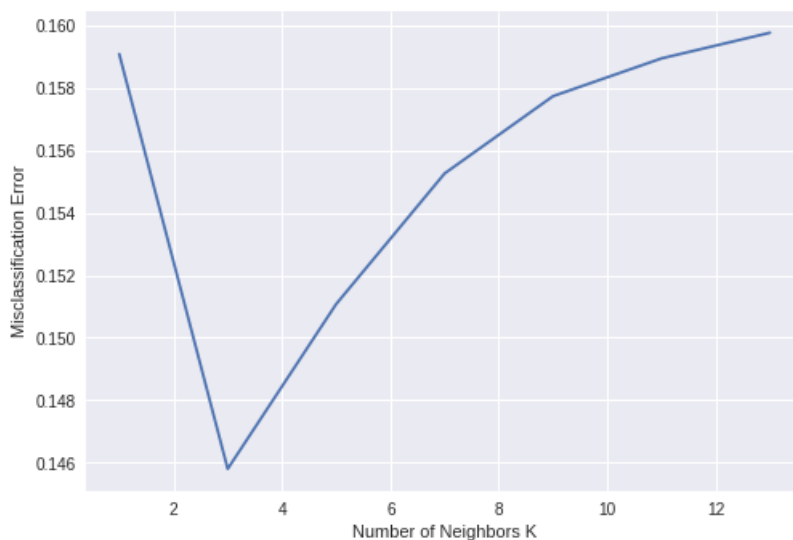
The f1 score of the knn classifier of BOW for k = 3 is 88.333333%

In [12]:

```
import matplotlib.pyplot as plt
import numpy as np
# plot misclassification error vs k
plt.plot(neighbors, MSE)

plt.xlabel('Number of Neighbors K')
plt.ylabel('Misclassification Error')
plt.show()

print("the misclassification error for each k value is : ", np.round(MSE,3))
```



the misclassification error for each k value is : [0.159 0.146 0.151 0.155 0.158 0.159 0.16 ]

In [0]:

## TFIDF

In [0]:

```
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer
```

```

X_tra, X_tes, y_train, y_test = train_test_split(x.values,y.values,test_size=0.3,shuffle=False)

#Implementing BAG of words
tfidf = TfidfVectorizer(ngram_range=(0,1),dtype=float)
X_tf =tfidf.fit_transform(X_tra)

# Standerdising the data
norm = StandardScaler(with_mean = False)
X_train = norm.fit_transform(X_tf)

# tfidf test
X_tfte = tfidf.transform(X_tes)

# Standerdising the data
X_test = norm.transform(X_tfte)

```

In [14]:

```

# 10 fold Cross Validation
myList = list(range(0,50))
neighbors = list(filter(lambda x: x % 2 != 0, myList))# Finding prime numbers

cv_scores = []

for k in neighbors:
    knn = KNeighborsClassifier(n_neighbors=k,algorithm='brute')
    scores = cross_val_score(knn, X_train,y_train, cv=10, scoring='f1_weighted')
    print('\nThe cross validation score for K = {} is {}'.format(k,scores.mean()))
    cv_scores.append(scores.mean())

MSE = [1 - x for x in cv_scores]

optimal_k = neighbors[MSE.index(min(MSE))]

```

The cross validation score for K = 1 is 0.8419452942163795.

The cross validation score for K = 3 is 0.8450248777514654.

The cross validation score for K = 5 is 0.8414814566581768.

```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)

```

The cross validation score for K = 7 is 0.8400123163639247.

```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
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/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)

```

The cross validation score for K = 9 is 0.8393238274453758.

```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:

```

```

UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
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mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)

```

The cross validation score for K = 11 is 0.8388895100916157.

```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
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'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
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'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
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/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
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mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)

```

The cross validation score for K = 13 is 0.8388024646892724.

```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa

```





```
mple.  
'precision', 'predicted', average, warn_for)  
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:  
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa  
mple.  
'precision', 'predicted', average, warn_for)
```

The cross validation score for  $K = 17$  is 0.8387154192869289.

[illegible]

The cross validation score for  $K = 19$  is 0.8387154192869289.

```
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:  
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa  
mples.  
    'precision', 'predicted', average, warn_for)  
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:  
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa  
mples.  
    'precision', 'predicted', average, warn_for)  
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:  
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa  
mples.  
    'precision', 'predicted', average, warn_for)  
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:  
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mples.  
    'precision', 'predicted', average, warn_for)  
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:  
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa  
mples.
```

```
mples.  
  'precision', 'predicted', average, warn_for)  
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:  
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa  
mples.  
  'precision', 'predicted', average, warn_for)  
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:  
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa  
mples.  
  'precision', 'predicted', average, warn_for)  
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:  
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mples.  
  'precision', 'predicted', average, warn_for)  
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:  
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa  
mples.  
  'precision', 'predicted', average, warn_for)
```

The cross validation score for  $K = 21$  is 0.8387154192869289.

[illegible]

The cross validation score for  $K = 23$  is 0.8387154192869289.

```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
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/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
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  'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.

```

[illegible]

The cross validation score for  $K = 25$  is 0.8387154192869289.

[illegible]

The cross validation score for  $K = 27$  is 0.8387154192869289.



```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
    'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
    'precision', 'predicted', average, warn_for)

```

The cross validation score for  $K = 31$  is 0.8387154192869289.

[illegible]

The cross validation score for K = 33 is 0.8387154192869289.

```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
    'precision', 'predicted', average, warn_for)
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```

```

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mples.
'precision', 'predicted', average, warn_for)

```

The cross validation score for K = 35 is 0.8387154192869289.

```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)

```

The cross validation score for K = 37 is 0.8387154192869289.

```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)

```





The cross validation score for  $K = 41$  is 0.8387154192869289.

[illegible]

The cross validation score for  $K = 43$  is 0.8387154192869289.

[illegible]





```

mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)

```

The cross validation score for K = 49 is 0.8387154192869289.

```

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135:
UndefinedMetricWarning: F-score is ill-defined and being set to 0.0 in labels with no predicted sa
mples.
'precision', 'predicted', average, warn_for)

```

In [15]:

```
print('\nThe optimal number of neighbors is %d.' % optimal_k)
```

The optimal number of neighbors is 3.

In [0]:

```

knn_optimal = KNeighborsClassifier(n_neighbors=optimal_k,algorithm='brute')

knn_optimal.fit(X_train,y_train)

pred = knn_optimal.predict(X_test)

```

In [17]:

```

f1 = f1_score(y_test, pred, average='micro')
f1_per= f1*100

print('\nThe f1 score of the knn classifier of BOW for k = %d is %f%' % (optimal_k, f1_per))

```

The f1 score of the knn classifier of BOW for k = 3 is 88.425000%

In [18]:

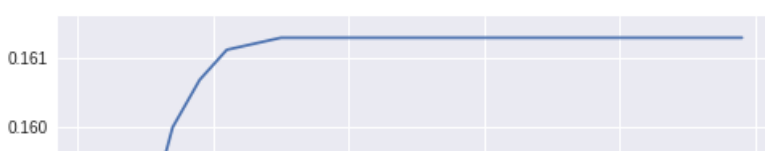
```

import matplotlib.pyplot as plt
import numpy as np
# plot misclassification error vs k
plt.plot(neighbors, MSE)

plt.xlabel('Number of Neighbors K')
plt.ylabel('Misclassification Error')
plt.show()

print("the misclassification error for each k value is : ", np.round(MSE,3))

```





```

Collecting boto>=2.3.2 (from smart-open>=1.2.1->gensim)
  Downloading
https://files.pythonhosted.org/packages/23/10/c0b78c27298029e4454a472a1919bde20cb182dab1662cec7f2ca523/boto-2.49.0-py2.py3-none-any.whl (1.4MB)
100% |████████████████████████████████████████| 1.4MB 15.1MB/s
Collecting bz2file (from smart-open>=1.2.1->gensim)
  Downloading
https://files.pythonhosted.org/packages/61/39/122222b5e85cd41c391b68a99ee296584b2a2d1d233e7ee32b453f2d/bz2file-0.98.tar.gz
Requirement already satisfied: requests in /usr/local/lib/python3.6/dist-packages (from smart-open>=1.2.1->gensim) (2.18.4)
Collecting boto3 (from smart-open>=1.2.1->gensim)
  Downloading
https://files.pythonhosted.org/packages/a0/d8/8b000ffeba218d47cd81fbd0bd0b2790742f81ffe116964a298be8a4/boto3-1.9.50-py2.py3-none-any.whl (128kB)
100% |████████████████████████████████████████| 133kB 26.9MB/s
Requirement already satisfied: idna<2.7,>=2.5 in /usr/local/lib/python3.6/dist-packages (from requests->smart-open>=1.2.1->gensim) (2.6)
Requirement already satisfied: chardet<3.1.0,>=3.0.2 in /usr/local/lib/python3.6/dist-packages (from requests->smart-open>=1.2.1->gensim) (3.0.4)
Requirement already satisfied: urllib3<1.23,>=1.21.1 in /usr/local/lib/python3.6/dist-packages (from requests->smart-open>=1.2.1->gensim) (1.22)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.6/dist-packages (from requests->smart-open>=1.2.1->gensim) (2018.10.15)
Collecting jmespath<1.0.0,>=0.7.1 (from boto3->smart-open>=1.2.1->gensim)
  Downloading
https://files.pythonhosted.org/packages/b7/31/05c8d001f7f87f0f07289a5fc0fc3832e9a57f2dbd4d3b0fee70e365/jmespath-0.9.3-py2.py3-none-any.whl
Collecting botocore<1.13.0,>=1.12.50 (from boto3->smart-open>=1.2.1->gensim)
  Downloading
https://files.pythonhosted.org/packages/82/c1/a26012b4dbca2e2ae06d7b24dc6b4378c0b0544d25f3e3f216612033/botocore-1.12.50-py2.py3-none-any.whl (4.9MB)
100% |████████████████████████████████████████| 4.9MB 6.8MB/s
Collecting s3transfer<0.2.0,>=0.1.10 (from boto3->smart-open>=1.2.1->gensim)
  Downloading
https://files.pythonhosted.org/packages/d7/14/2a0004d487464d120c9fb85313a75cd3d71a7506955be458eebfcbld/s3transfer-0.1.13-py2.py3-none-any.whl (59kB)
100% |████████████████████████████████████████| 61kB 23.1MB/s
Requirement already satisfied: python-dateutil<3.0.0,>=2.1; python_version >= "2.7" in /usr/local/lib/python3.6/dist-packages (from botocore<1.13.0,>=1.12.50->boto3->smart-open>=1.2.1->gensim) (2.5.3)
Collecting docutils>=0.10 (from botocore<1.13.0,>=1.12.50->boto3->smart-open>=1.2.1->gensim)
  Downloading
https://files.pythonhosted.org/packages/36/fa/08e9e6e0e3cbd1d362c3bbee8d01d0aedb2155c4ac112b19ef3cad8d/docutils-0.14-py3-none-any.whl (543kB)
100% |████████████████████████████████████████| 552kB 24.1MB/s
Building wheels for collected packages: smart-open, bz2file
  Running setup.py bdist_wheel for smart-open ... - \ done
  Stored in directory:
/root/.cache/pip/wheels/23/00/44/e5b939f7a80c04e32297dbd6d96fa3065af89ecf57e2b5f89f
  Running setup.py bdist_wheel for bz2file ... - done
  Stored in directory:
/root/.cache/pip/wheels/81/75/d6/e1317bf09b1af5a30befc2a007869fa6e1f516b8f7c591cb9
Successfully built smart-open bz2file
Installing collected packages: boto, bz2file, jmespath, docutils, botocore, s3transfer, boto3, smart-open, gensim
Successfully installed boto-2.49.0 boto3-1.9.50 botocore-1.12.50 bz2file-0.98 docutils-0.14 gensim-3.6.0 jmespath-0.9.3 s3transfer-0.1.13 smart-open-1.7.1

```

In [0]:

```

pickle.dump(y_train, open('y_train.p', 'wb'))
pickle.dump(y_test, open('y_test.p', 'wb'))

```

In [23]:

```

#word to vector
from gensim.models import Word2Vec
w2v_model=Word2Vec(sent_of_train,min_count=3,size=200, workers=4)# words which occurs 3 times; 500 dimensions

w2v_words = list(w2v_model.wv.vocab)
print("number of words that occurred minimum 3 times ",len(w2v_words))

```

number of words that occurred minimum 3 times 10559

In [24]:

```
# compute average word2vec for each review for X_train .
from tqdm import tqdm
import numpy as np

train_vectors = []
for sent in tqdm(sent_of_test):
    sent_vec = np.zeros(200)
    cnt_words = 0;
    for word in sent:
        if word in w2v_words:
            vec = w2v_model.wv[word]
            sent_vec += vec
            cnt_words += 1
    if cnt_words != 0:
        sent_vec /= cnt_words
    train_vectors.append(sent_vec)

train_vectors1 = []
for sent in tqdm(sent_of_train):
    sent_vec = np.zeros(200)
    cnt_words = 0;
    for word in sent:
        if word in w2v_words:
            vec = w2v_model.wv[word]
            sent_vec += vec
            cnt_words += 1
    if cnt_words != 0:
        sent_vec /= cnt_words
    train_vectors1.append(sent_vec)

len(train_vectors), len(train_vectors1)
```

```
100%|██████████| 15000/15000 [00:25<00:00, 588.50it/s]
100%|██████████| 35000/35000 [00:55<00:00, 629.57it/s]
```

Out[24]:

```
(15000, 35000)
```

In [25]:

```
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import TimeSeriesSplit

# Data-preprocessing: Standardizing the data
sc = StandardScaler(with_mean = False)
X_train3 = sc.fit_transform(train_vectors1)
X_test3 = sc.transform(train_vectors)

y_train.shape, X_train3.shape, X_test3.shape, y_test.shape
```

Out[25]:

```
((35000,), (35000, 200), (15000, 200), (15000,))
```

In [0]:

```
#####
X_train3 = pickle.load(open('drive/My Drive/Colab Notebooks/KNN/X_train3.p', 'rb'))
X_test3 = pickle.load(open('drive/My Drive/Colab Notebooks/KNN/X_test3.p', 'rb'))
```

In [0]:

In [26]:

```
X_train3.shape,X_test3.shape, y_train.shape, y_test.shape
```

Out[26]:

```
((35000, 200), (15000, 200), (35000,), (15000,))
```

In [27]:

```
#####

from sklearn.neighbors import KNeighborsClassifier
from sklearn.cross_validation import cross_val_score
from sklearn.metrics import make_scorer
from sklearn.metrics import f1_score

# 10 fold Cross Validation
myList = list(range(0,15))
neighbors = list(filter(lambda x: x % 2 != 0, myList))

cv_scores = []

for k in neighbors:
    knn = KNeighborsClassifier(n_neighbors=k,algorithm='brute')
    scores = cross_val_score(knn,X_train3,y_train, cv=10, scoring='f1_weighted')
    print('\nThe cross validation score for K = {} is {}'.format(k,scores.mean()))
    cv_scores.append(scores.mean())

MSE = [1 - x for x in cv_scores]

optimal_k = neighbors[MSE.index(min(MSE))]
```

The cross validation score for K = 1 is 0.85807776790818.

The cross validation score for K = 3 is 0.8729255886114174.

The cross validation score for K = 5 is 0.8755752538622973.

The cross validation score for K = 7 is 0.8749558733332183.

The cross validation score for K = 9 is 0.8739550089642041.

The cross validation score for K = 11 is 0.8731956450497907.

The cross validation score for K = 13 is 0.8722109996489932.

In [28]:

```
print('\nThe optimal number of neighbors is %d.' % optimal_k)
```

The optimal number of neighbors is 5.

In [0]:

```
from sklearn.neighbors import KNeighborsClassifier
knn_optimal = KNeighborsClassifier(n_neighbors=5,algorithm='brute')
```

In [0]:

```
knn_optimal.fit(X_train3,y_train)

pred = knn_optimal.predict(X_test3)
```

In [32]:

```
f1 = f1_score(y_test, pred, average='micro')
```

```
f1_sc =f1*100

print('\nThe f1 score of the knn classifier of BOW for k = %d is %f%%' % (5, f1_sc))
```

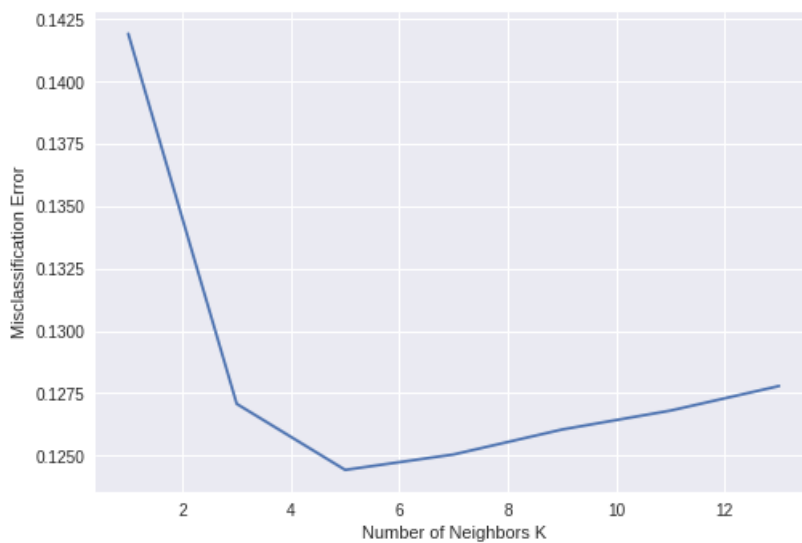
The f1 score of the knn classifier of BOW for k = 5 is 89.000000%

In [33]:

```
import matplotlib.pyplot as plt
import numpy as np
# plot misclassification error vs k
plt.plot(neighbors, MSE)

plt.xlabel('Number of Neighbors K')
plt.ylabel('Misclassification Error')
plt.show()

print("the misclassification error for each k value is : ", np.round(MSE,3))
```



the misclassification error for each k value is : [0.142 0.127 0.124 0.125 0.126 0.127 0.128]

## TFIDF-Word2vec

In [35]:

```
y = final['Score'].iloc[:50000]
x = final['CleanedText'].iloc[:50000]
x.shape,y.shape
```

Out[35]:

((50000,), (50000,))

In [0]:

```
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer

X_tra, X_tes, y_train, y_test = train_test_split(x.values,y.values,test_size=0.3,random_state=0)
```

In [0]:

```
sent_of_train=[]
for sent in X_tra:
    sent of train.append(sent.split())
```

```
sent_of_test=[]
for sent in X_tes:
    sent_of_test.append(sent.split())
```

In [38]:

```
#word to vector
from gensim.models import Word2Vec
w2v_model=Word2Vec(sent_of_train,min_count=3,size=200, workers=4)# words which occurs 3 times; 500
dimensions

w2v_words = list(w2v_model.wv.vocab)
print("number of words that occurred minimum 3 times ",len(w2v_words))
```

number of words that occurred minimum 3 times 10497

In [0]:

```
m = TfidfVectorizer()
tf_idf_matrix = m.fit_transform(X_tra)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(m.get_feature_names(), list(m.idf_)))
```

In [40]:

```
from tqdm import tqdm
import numpy as np

# TF-IDF weighted Word2Vec
tfidf_feat = m.get_feature_names() # tfidf words/col-names
# final_tf_idf is the sparse matrix with row= sentence, col=word and cell_val = tfidf

tfidf_sent_vectors = []; # the tfidf-w2v for each sentence/review is stored in this list
row=0;
for sent in tqdm(sent_of_train): # for each review/sentence
    sent_vec = np.zeros(200) # as word vectors are of zero length
    weight_sum =0; # num of words with a valid vector in the sentence/review
    for word in sent: # for each word in a review/sentence
        if word in w2v_words:
            vec = w2v_model.wv[word]
            tf_idf = dictionary[word]*(sent.count(word)/len(sent))
            sent_vec += (vec * tf_idf)
            weight_sum += tf_idf
    if weight_sum != 0:
        sent_vec /= weight_sum
    tfidf_sent_vectors.append(sent_vec)
    row += 1
```

100%|██████████| 35000/35000 [01:02<00:00, 555.88it/s]

In [41]:

```
tfidf_sent_vectors1 = []; # the tfidf-w2v for each sentence/review is stored in this list
row=0;
for sent in tqdm(sent_of_test): # for each review/sentence
    sent_vec = np.zeros(200) # as word vectors are of zero length
    weight_sum =0; # num of words with a valid vector in the sentence/review
    for word in sent: # for each word in a review/sentence
        if word in w2v_words:
            vec = w2v_model.wv[word]
            tf_idf = dictionary[word]*(sent.count(word)/len(sent))
            sent_vec += (vec * tf_idf)
            weight_sum += tf_idf
    if weight_sum != 0:
        sent_vec /= weight_sum
    tfidf_sent_vectors1.append(sent_vec)
    row += 1
```

100%|██████████| 15000/15000 [00:30<00:00, 495.73it/s]



In [0]:

```
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import TimeSeriesSplit

# Data-preprocessing: Standardizing the data
sc = StandardScaler(with_mean = False)
X_train4 = sc.fit_transform(tfidf_sent_vectors)
X_test4 = sc.transform(tfidf_sent_vectors1)
```

In [43]:

```
X_train4.shape,X_test4.shape
```

Out[43]:

```
((35000, 200), (15000, 200))
```

In [0]:

```
import pickle

pickle.dump(X_train4, open('X_train4.p', 'wb'))
pickle.dump(X_test4, open('X_test4.p', 'wb'))
```

In [44]:

```
X_train4.shape,X_test4.shape, y_train.shape, y_test.shape
```

Out[44]:

```
((35000, 200), (15000, 200), (35000,), (15000,))
```

In [45]:

```
#####

from sklearn.neighbors import KNeighborsClassifier
from sklearn.cross_validation import cross_val_score
from sklearn.metrics import make_scorer
from sklearn.metrics import f1_score

# 10 fold Cross Validation
myList = list(range(0,15))
neighbors = list(filter(lambda x: x % 2 != 0, myList))

cv_scores = []

for k in neighbors:
    knn = KNeighborsClassifier(n_neighbors=k,algorithm='brute')
    scores = cross_val_score(knn,X_train4,y_train, cv=10, scoring='f1_weighted')
    print('\nThe cross validation score for K = {} is {}'.format(k,scores.mean()))
    cv_scores.append(scores.mean())

MSE = [1 - x for x in cv_scores]

optimal_k = neighbors[MSE.index(min(MSE))]
```

The cross validation score for K = 1 is 0.8479195225642234.

The cross validation score for K = 3 is 0.8637196918275543.

The cross validation score for K = 5 is 0.8662677218544383.

The cross validation score for K = 7 is 0.8657491941284302.

The cross validation score for K = 9 is 0.8651090093086712.

The cross validation score for K = 11 is 0.8630570690178787

The cross validation score for K = 11 is 0.85557889017877.

The cross validation score for K = 13 is 0.8610161430560955.

In [46]:

```
print('\nThe optimal number of neighbors is %d.' % optimal_k)
```

The optimal number of neighbors is 5.

In [0]:

```
x_t=np.array(X_train4)
y_t=np.array(y_train)
X_tes=np.array(X_test4)
```

In [0]:

```
knn_optimal = KNeighborsClassifier(n_neighbors=5,algorithm='brute')
knn_optimal.fit(X_train4,y_train)
pred = knn_optimal.predict(X_test4)
```

In [48]:

```
f1 = f1_score(y_test, pred, average='micro')
f1_sc =f1*100
print('\nThe f1 score of the knn classifier of BOW for k = %d is %f%%' % (5, f1_sc))
```

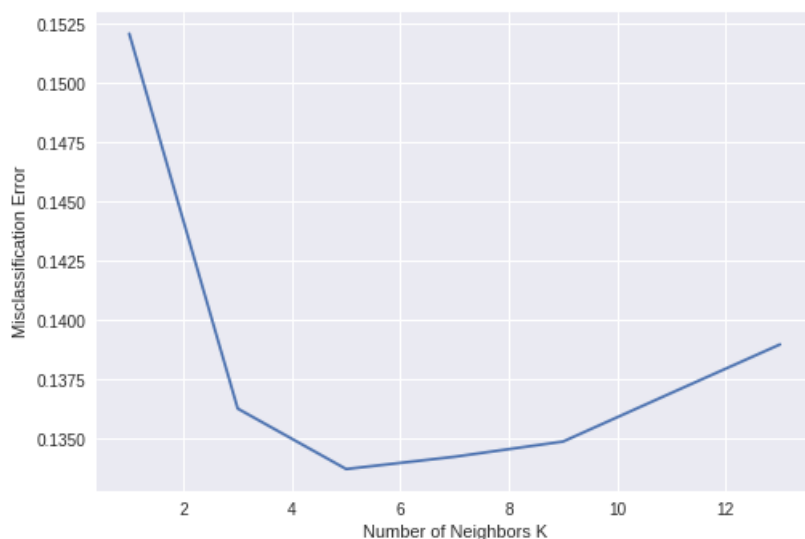
The f1 score of the knn classifier of BOW for k = 5 is 89.086667%

In [49]:

```
import matplotlib.pyplot as plt
import numpy as np
# plot misclassification error vs k
plt.plot(neighbors, MSE)

plt.xlabel('Number of Neighbors K')
plt.ylabel('Misclassification Error')
plt.show()

print("the misclassification error for each k value is : ", np.round(MSE,3))
```



the misclassification error for each k value is : [0.152 0.136 0.134 0.134 0.135 0.137 0.139]

- =====
- F1 score of BOW at hyperparameter K = 3

83.88%

- F1 score of TFIDF at hyperparameter K = 3

88.45%

- F1 score of AVG W2V at hyperparameter K=5

89%

- F1 score of TFIDF W2V at hyperparameter k=5

89.09%

=====

